```
import torch
import torchvision as tv
B = 100
transf = tv.transforms.Compose(
           [ tv.transforms.ToTensor().
             tv.transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.5, 0.5))
           1)
                                   # ( value - mean ) / std
trn data = tv.datasets.CIFAR10( root='./data', train=True,
                           download=True, transform=transf)
tst data = tv.datasets.CIFAR10( root='./data', train=False,
                           download=True, transform=transf)
trn load = torch.utils.data.DataLoader( trn data, batch size=B,
                                     shuffle=True, num workers=2)
tst load = torch.utils.data.DataLoader( tst data, batch size=B,
                                     shuffle=False, num workers=2)
classes = ('plane', 'car', 'bird', 'cat', 'deer',
           'dog', 'frog', 'horse', 'ship', 'truck')
C = len(classes)
# idata = iter( tst load)
# image, label = next( idata)
class ConvNet( torch.nn.Module):
    def init ( , num classes):
                                           \# 0 = | (I + 2p - k)/s + 1 |
        super(). init ()
                                           # Tambien pueden ser tuplas.
        .conv1 = torch.nn.Conv2d( 3, 16, kernel size=5,
                                               stride=2, padding=2)
        .conv2 = torch.nn.Conv2d( 16, 32, kernel size=3,
                                               stride=1, padding=1)
       .mpool = torch.nn.MaxPool2d( kernel size=2, stride=2)
        .hlin1 = torch.nn.Linear( 32*8*8, 512)
        .hlin2 = torch.nn.Linear( 512, num classes)
    def forward( , x):
                                               # [B, 3, 32, 32]
       h1 = .conv1(x).relu()
                                               # [B, 16, 16, 16]
       h2 = .mpool( .conv2( h1)).relu()
                                               # [B, 32, 8, 8]
       h2 = h2.view(-1, 32*8*8)
                                               # [B, 32 * 8 * 8]
       h3 = .hlin1(h2).tanh()
                                               # [B, 512]
       y = .hlin2(h3)
                                               # [B, 10]
        return y
```

```
device = torch.device('cuda:0' if torch.cuda.is available() else 'cpu')
T = 5
model = ConvNet( C).to( device)
optim = torch.optim.Adam( model.parameters())
costf = torch.nn.CrossEntropyLoss()
model.train()
for t in range(T):
    for i, (images, labels) in enumerate(trn load):
        images = images.to( device)
        labels = labels.to( device)
        output = model( images)
        error = costf( output, labels)
        optim.zero grad()
        error.backward()
        optim.step()
        if (i+1)%100 == 0:
            print ( "Epoch [{}/{}], Batch [{}], Loss: {:.4f}"
                    .format( t+1, T, i+1, error.item()))
model.eval()
c, t = 0, 0
with torch.no grad():
   correct = 0
    total = 0
    for images, labels in tst load:
        images = images.to( device)
        labels = labels.to( device)
        output = model( images)
        , predicted = torch.max( output.data, 1)
        c += (predicted == labels).sum().item()
       t += labels.size(0)
print( "Accuracy: {} %".format(100*c/t))
torch.save( model, 'convnet.model')
# .enc = nn.Sequential( nn.Conv2d(3, 6, kernel size=5), nn.ReLU(True),
                         nn.Conv2d(6,16, kernel size=5), nn.ReLU(True))
  .dec = nn.Sequential( nn.ConvTranspose2d(16,6, 5), nn.ReLU(True),
                         nn.ConvTranspose2d( 6,3, 5), nn.Sigmoid() )
```